

```
In [1]: 1 import pandas as pd
```

```
In [2]: 1 # 2 main datatypes  
2 series = pd.Series(['BMW', 'Toyota', 'Honda'])
```

```
In [3]: 1 series
```

```
Out[3]: 0      BMW  
1     Toyota  
2      Honda  
dtype: object
```

```
In [4]: 1 # series = 1-dimensional
```

```
In [5]: 1 colours = pd.Series(['red', 'blue', 'white'])  
2 colours
```

```
Out[5]: 0      red  
1     blue  
2    white  
dtype: object
```

```
In [6]: 1 # Dataframe = 2-dimentional  
2 car_data = pd.DataFrame({'car make': series, 'colours': colours})  
3 car_data
```

```
Out[6]:
```

	car make	colours
0	BMW	red
1	Toyota	blue
2	Honda	white

```
In [7]: 1 # import data
        2 car_sales = pd.read_csv('car-sales.csv')
        3 car_sales
```

Out[7]:

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
1	Honda	Red	87899	4	\$5,000.00
2	Toyota	Blue	32549	3	\$7,000.00
3	BMW	Black	11179	5	\$22,000.00
4	Nissan	White	213095	4	\$3,500.00
5	Toyota	Green	99213	4	\$4,500.00
6	Honda	Blue	45698	4	\$7,500.00
7	Honda	Blue	54738	4	\$7,000.00
8	Toyota	White	60000	4	\$6,250.00
9	Nissan	White	31600	4	\$9,700.00

```
In [8]: 1 # exporting a dataframe
        2 car_data.to_csv('exported-car-sales.csv', index = False)
```

```
In [9]: 1 exported_car_sales = pd.read_csv('exported-car-sales.csv')
        2 exported_car_sales
```

Out[9]:

	car make	colours
0	BMW	red
1	Toyota	blue
2	Honda	white

```
In [10]: 1 # 2 exporting a dataframe
        2 car_sales.to_csv('exported-2nd-car-sales.csv', index=False)
```

```
In [11]: 1 exported_2nd_car_sales = pd.read_csv('exported-2nd-car-sales.csv')
        2 exported_2nd_car_sales
```

Out[11]:

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
1	Honda	Red	87899	4	\$5,000.00
2	Toyota	Blue	32549	3	\$7,000.00
3	BMW	Black	11179	5	\$22,000.00
4	Nissan	White	213095	4	\$3,500.00
5	Toyota	Green	99213	4	\$4,500.00
6	Honda	Blue	45698	4	\$7,500.00
7	Honda	Blue	54738	4	\$7,000.00
8	Toyota	White	60000	4	\$6,250.00
9	Nissan	White	31600	4	\$9,700.00

## Describe Data

```
In [12]: 1 # Attributes (it has no bracket)
        2 car_sales.dtypes
        3
        4 # function
        5 # car_sales.to_csv(): it has a bracket
```

Out[12]:

Make	object
Colour	object
Odometer (KM)	int64
Doors	int64
Price	object
dtype:	object

```
In [13]: 1 car_sales.columns
```

```
Out[13]: Index(['Make', 'Colour', 'Odometer (KM)', 'Doors', 'Price'], dtype='object')
```

```
In [14]: 1 car_columns = car_sales.columns  
2 car_columns
```

```
Out[14]: Index(['Make', 'Colour', 'Odometer (KM)', 'Doors', 'Price'], dtype='object')
```

```
In [15]: 1 car_sales.index
```

```
Out[15]: RangeIndex(start=0, stop=10, step=1)
```

```
In [16]: 1 car_sales
```

```
Out[16]:
```

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
1	Honda	Red	87899	4	\$5,000.00
2	Toyota	Blue	32549	3	\$7,000.00
3	BMW	Black	11179	5	\$22,000.00
4	Nissan	White	213095	4	\$3,500.00
5	Toyota	Green	99213	4	\$4,500.00
6	Honda	Blue	45698	4	\$7,500.00
7	Honda	Blue	54738	4	\$7,000.00
8	Toyota	White	60000	4	\$6,250.00
9	Nissan	White	31600	4	\$9,700.00

In [17]: 1 car\_sales.describe

Out[17]: <bound method NDFrame.describe of

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
1	Honda	Red	87899	4	\$5,000.00
2	Toyota	Blue	32549	3	\$7,000.00
3	BMW	Black	11179	5	\$22,000.00
4	Nissan	White	213095	4	\$3,500.00
5	Toyota	Green	99213	4	\$4,500.00
6	Honda	Blue	45698	4	\$7,500.00
7	Honda	Blue	54738	4	\$7,000.00
8	Toyota	White	60000	4	\$6,250.00
9	Nissan	White	31600	4	\$9,700.00

In [18]: 1 *# functions: it has bracket*  
2 car\_sales.describe()

Out[18]:

	Odometer (KM)	Doors
<b>count</b>	10.000000	10.000000
<b>mean</b>	78601.400000	4.000000
<b>std</b>	61983.471735	0.471405
<b>min</b>	11179.000000	3.000000
<b>25%</b>	35836.250000	4.000000
<b>50%</b>	57369.000000	4.000000
<b>75%</b>	96384.500000	4.000000
<b>max</b>	213095.000000	5.000000

In [19]: 1 car\_sales.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10 entries, 0 to 9
Data columns (total 5 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Make            10 non-null    object
1   Colour          10 non-null    object
2   Odometer (KM)    10 non-null    int64
3   Doors           10 non-null    int64
4   Price           10 non-null    object
dtypes: int64(2), object(3)
memory usage: 528.0+ bytes
```

In [20]: 1 car\_sales.mean()

C:\Users\USER\AppData\Local\Temp\ipykernel\_5644\4073448239.py:1: FutureWarning: Dropping of nuisance columns in DataFrame reductions (with 'numeric\_only=None') is deprecated; in a future version this will raise TypeError. Select only valid columns before calling the reduction.

```
car_sales.mean()
```

Out[20]: Odometer (KM) 78601.4  
Doors 4.0  
dtype: float64

In [21]: 1 car\_prices = pd.Series([3000, 1500, 111250])  
2 car\_prices.mean()

Out[21]: 38583.333333333336

In [22]: 1 car\_sales.sum()

Out[22]: Make ToyotaHondaToyotaBMWNissanToyotaHondaHondaToyo...  
Colour WhiteRedBlueBlackWhiteGreenBlueBlueWhiteWhite  
Odometer (KM) 786014  
Doors 40  
Price \$4,000.00 \$5,000.00 \$7,000.00 \$22,000.00 \$3,50...  
dtype: object

```
In [23]: 1 car_sales['Doors'].sum()
```

```
Out[23]: 40
```

```
In [24]: 1 len(car_sales)
```

```
Out[24]: 10
```

```
In [25]: 1 car_sales
```

```
Out[25]:
```

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
1	Honda	Red	87899	4	\$5,000.00
2	Toyota	Blue	32549	3	\$7,000.00
3	BMW	Black	11179	5	\$22,000.00
4	Nissan	White	213095	4	\$3,500.00
5	Toyota	Green	99213	4	\$4,500.00
6	Honda	Blue	45698	4	\$7,500.00
7	Honda	Blue	54738	4	\$7,000.00
8	Toyota	White	60000	4	\$6,250.00
9	Nissan	White	31600	4	\$9,700.00

## viewing and selecting data

In [26]:

```
1 car_sales.head()
```

Out[26]:

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
1	Honda	Red	87899	4	\$5,000.00
2	Toyota	Blue	32549	3	\$7,000.00
3	BMW	Black	11179	5	\$22,000.00
4	Nissan	White	213095	4	\$3,500.00

In [27]:

```
1 car_sales
```

Out[27]:

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
1	Honda	Red	87899	4	\$5,000.00
2	Toyota	Blue	32549	3	\$7,000.00
3	BMW	Black	11179	5	\$22,000.00
4	Nissan	White	213095	4	\$3,500.00
5	Toyota	Green	99213	4	\$4,500.00
6	Honda	Blue	45698	4	\$7,500.00
7	Honda	Blue	54738	4	\$7,000.00
8	Toyota	White	60000	4	\$6,250.00
9	Nissan	White	31600	4	\$9,700.00



```
In [28]: 1 car_sales.head(7)
```

Out[28]:

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
1	Honda	Red	87899	4	\$5,000.00
2	Toyota	Blue	32549	3	\$7,000.00
3	BMW	Black	11179	5	\$22,000.00
4	Nissan	White	213095	4	\$3,500.00
5	Toyota	Green	99213	4	\$4,500.00
6	Honda	Blue	45698	4	\$7,500.00

```
In [29]: 1 car_sales.tail()
```

Out[29]:

	Make	Colour	Odometer (KM)	Doors	Price
5	Toyota	Green	99213	4	\$4,500.00
6	Honda	Blue	45698	4	\$7,500.00
7	Honda	Blue	54738	4	\$7,000.00
8	Toyota	White	60000	4	\$6,250.00
9	Nissan	White	31600	4	\$9,700.00

```
In [30]: 1 car_sales.tail(3)
```

Out[30]:

	Make	Colour	Odometer (KM)	Doors	Price
7	Honda	Blue	54738	4	\$7,000.00
8	Toyota	White	60000	4	\$6,250.00
9	Nissan	White	31600	4	\$9,700.00

```
In [31]: 1 # .loc & .iloc
          2 animals = pd.Series(['cat', 'dog', 'pandas', 'snake'])
          3 animals
```

```
Out[31]: 0      cat
          1      dog
          2    pandas
          3     snake
          dtype: object
```

```
In [32]: 1 animals = pd.Series(['cat', 'dog', 'pandas', 'snake', 'birds'],
          2                        index = ([0, 3, 9, 8, 3]))
          3 animals
```

```
Out[32]: 0      cat
          3      dog
          9    pandas
          8     snake
          3     birds
          dtype: object
```

```
In [33]: 1 animals.loc[3]
```

```
Out[33]: 3      dog
          3     birds
          dtype: object
```

```
In [34]: 1 animals.loc[9]
```

```
Out[34]: 'pandas'
```

```
In [35]: 1 car_sales.loc[3]
```

```
Out[35]: Make      BMW
          Colour    Black
          Odometer (KM)  11179
          Doors      5
          Price      $22,000.00
          Name: 3, dtype: object
```

```
In [36]: 1 # .iloc refers to position  
2 animals.iloc[3]
```

Out[36]: 'snake'

```
In [37]: 1 animals
```

Out[37]: 0 cat  
3 dog  
9 pandas  
8 snake  
3 birds  
dtype: object

```
In [38]: 1 car_sales.iloc[3]
```

Out[38]: Make BMW  
Colour Black  
Odometer (KM) 11179  
Doors 5  
Price \$22,000.00  
Name: 3, dtype: object

```
In [39]: 1 car_sales
```

```
Out[39]:
```

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
1	Honda	Red	87899	4	\$5,000.00
2	Toyota	Blue	32549	3	\$7,000.00
3	BMW	Black	11179	5	\$22,000.00
4	Nissan	White	213095	4	\$3,500.00
5	Toyota	Green	99213	4	\$4,500.00
6	Honda	Blue	45698	4	\$7,500.00
7	Honda	Blue	54738	4	\$7,000.00
8	Toyota	White	60000	4	\$6,250.00
9	Nissan	White	31600	4	\$9,700.00

```
In [40]: 1 animals
```

```
Out[40]: 0      cat
3      dog
9  pandas
8   snake
3   birds
dtype: object
```

```
In [41]: 1 animals.iloc[:3]
```

```
Out[41]: 0      cat
3      dog
9  pandas
dtype: object
```

```
In [42]: 1 car_sales.loc[:3]
```

```
Out[42]:
```

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
1	Honda	Red	87899	4	\$5,000.00
2	Toyota	Blue	32549	3	\$7,000.00
3	BMW	Black	11179	5	\$22,000.00

```
In [43]: 1 car_sales.head(4)
```

```
Out[43]:
```

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
1	Honda	Red	87899	4	\$5,000.00
2	Toyota	Blue	32549	3	\$7,000.00
3	BMW	Black	11179	5	\$22,000.00

```
In [44]: 1 car_sales['Make']
```

```
Out[44]: 0    Toyota
1    Honda
2    Toyota
3    BMW
4    Nissan
5    Toyota
6    Honda
7    Honda
8    Toyota
9    Nissan
Name: Make, dtype: object
```

```
In [45]: 1 car_sales['Colour']
```

```
Out[45]: 0    White
          1     Red
          2    Blue
          3   Black
          4    White
          5   Green
          6    Blue
          7    Blue
          8    White
          9    White
          Name: Colour, dtype: object
```

```
In [46]: 1 car_sales['Make']
```

```
Out[46]: 0    Toyota
          1    Honda
          2    Toyota
          3     BMW
          4    Nissan
          5    Toyota
          6    Honda
          7    Honda
          8    Toyota
          9    Nissan
          Name: Make, dtype: object
```

```
In [47]: 1 car_sales.Make
```

```
Out[47]: 0    Toyota
          1    Honda
          2    Toyota
          3     BMW
          4    Nissan
          5    Toyota
          6    Honda
          7    Honda
          8    Toyota
          9    Nissan
          Name: Make, dtype: object
```

```
In [48]: 1 car_sales['Odometer (KM)']
```

```
Out[48]: 0    150043
         1     87899
         2     32549
         3     11179
         4    213095
         5     99213
         6     45698
         7     54738
         8     60000
         9     31600
         Name: Odometer (KM), dtype: int64
```

```
In [49]: 1 car_sales
```

```
Out[49]:
```

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
1	Honda	Red	87899	4	\$5,000.00
2	Toyota	Blue	32549	3	\$7,000.00
3	BMW	Black	11179	5	\$22,000.00
4	Nissan	White	213095	4	\$3,500.00
5	Toyota	Green	99213	4	\$4,500.00
6	Honda	Blue	45698	4	\$7,500.00
7	Honda	Blue	54738	4	\$7,000.00
8	Toyota	White	60000	4	\$6,250.00
9	Nissan	White	31600	4	\$9,700.00

In [50]: 1 car\_sales[car\_sales['Make']== 'Toyota']

Out[50]:

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
2	Toyota	Blue	32549	3	\$7,000.00
5	Toyota	Green	99213	4	\$4,500.00
8	Toyota	White	60000	4	\$6,250.00

In [51]: 1 car\_sales[car\_sales['Odometer (KM)'] > 100000]

Out[51]:

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
4	Nissan	White	213095	4	\$3,500.00

In [52]: 1 car\_sales[car\_sales['Odometer (KM)'] < 100000]

Out[52]:

	Make	Colour	Odometer (KM)	Doors	Price
1	Honda	Red	87899	4	\$5,000.00
2	Toyota	Blue	32549	3	\$7,000.00
3	BMW	Black	11179	5	\$22,000.00
5	Toyota	Green	99213	4	\$4,500.00
6	Honda	Blue	45698	4	\$7,500.00
7	Honda	Blue	54738	4	\$7,000.00
8	Toyota	White	60000	4	\$6,250.00
9	Nissan	White	31600	4	\$9,700.00



```
In [53]: 1 # comparing two columns
          2 pd.crosstab(car_sales['Make'], car_sales['Doors'])
          3
```

Out[53]:

	Doors	3	4	5
Make				
BMW	0	0	1	
Honda	0	3	0	
Nissan	0	2	0	
Toyota	1	3	0	

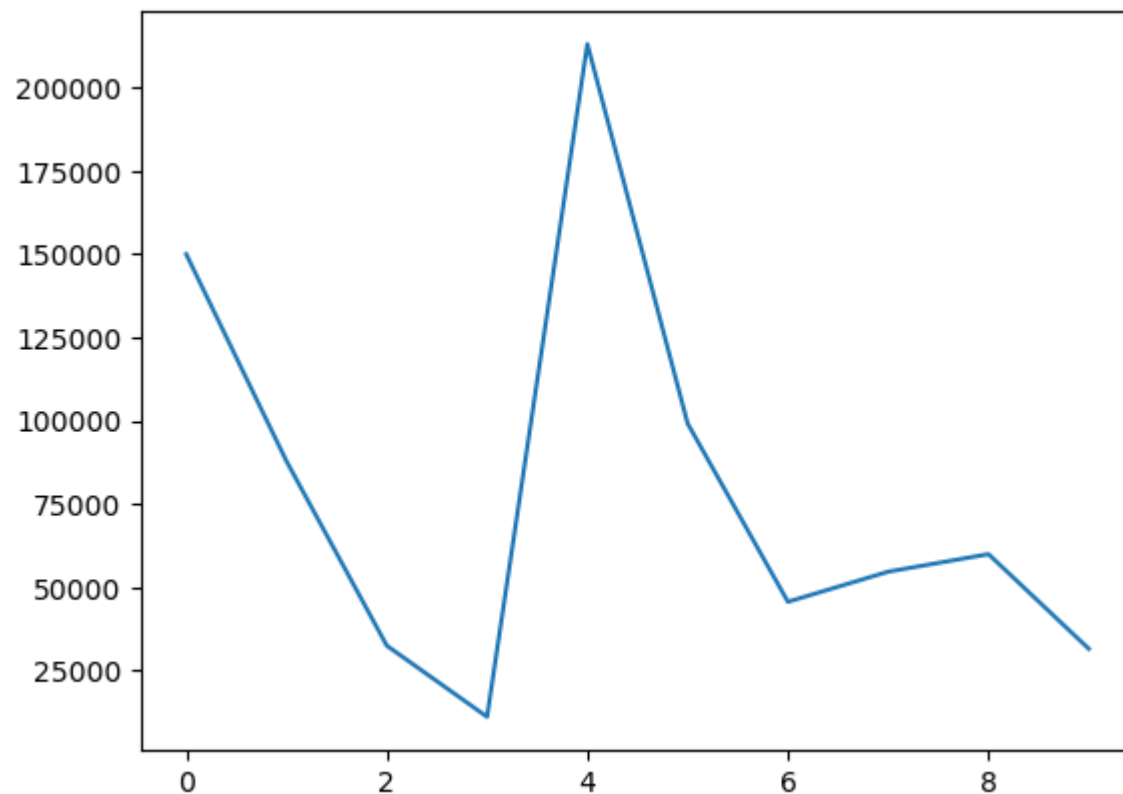
```
In [54]: 1 # comparing more columns: Groupby
          2 car_sales.groupby(['Make']).mean()
```

Out[54]:

	Odometer (KM)	Doors
Make		
BMW	11179.000000	5.00
Honda	62778.333333	4.00
Nissan	122347.500000	4.00
Toyota	85451.250000	3.75

```
In [64]: 1 car_sales['Odometer (KM)'].plot()
```

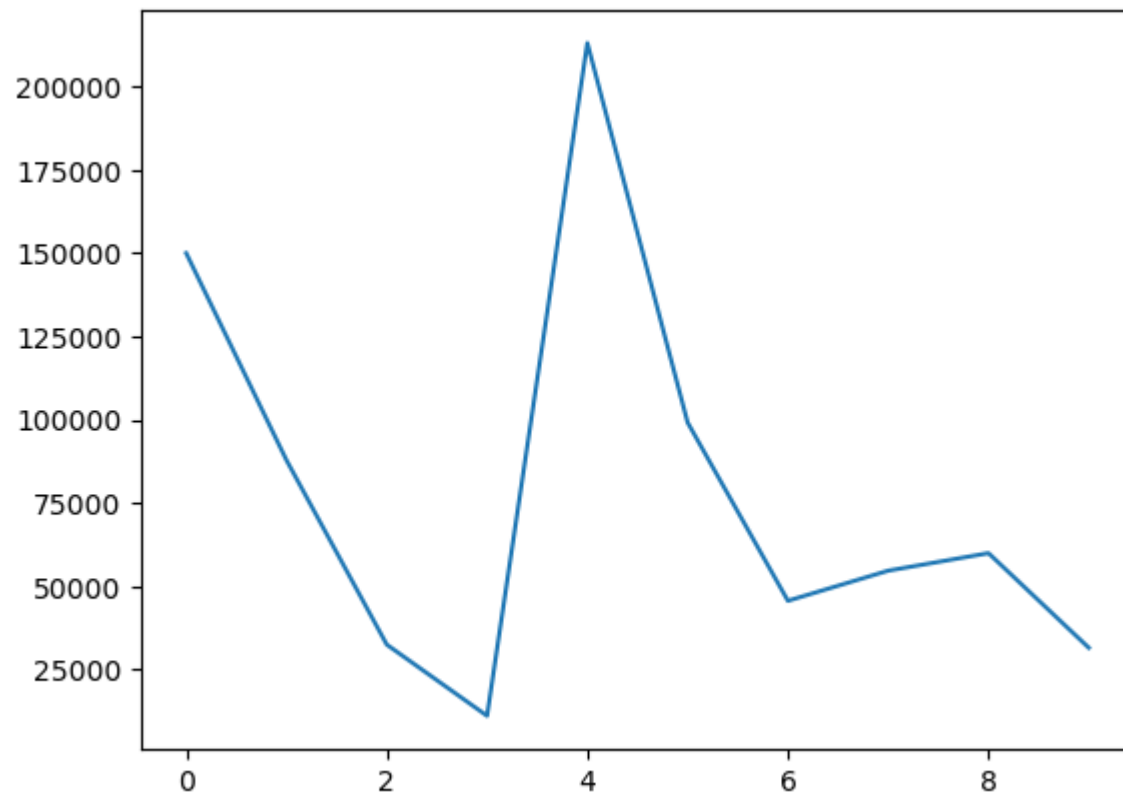
```
Out[64]: <AxesSubplot:>
```



```
In [65]: 1 # if the above code didn't run, we will use this to import matplotlib
2
3 %matplotlib inline
4 import matplotlib.pyplot as plt
```

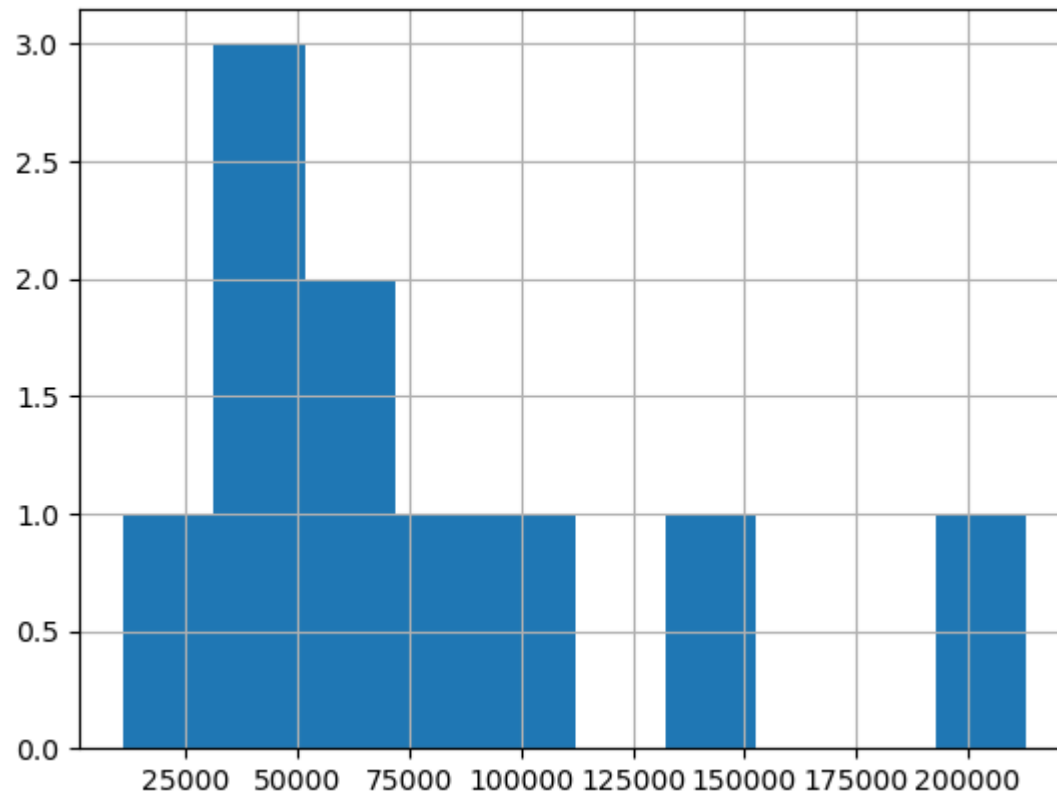
```
In [66]: 1 car_sales['Odometer (KM)'].plot()
```

```
Out[66]: <AxesSubplot:>
```



```
In [67]: 1 # histogram plot  
2 car_sales['Odometer (KM)'].hist()
```

Out[67]: <AxesSubplot:>



```
In [68]: 1 car_sales['Price'].dtype
```

Out[68]: dtype('O')

```
In [69]: 1 car_sales['Price'].plot()  
2 # price can not run cos is not a numerical data
```

```
-----  
TypeError                                Traceback (most recent call last)  
~\AppData\Local\Temp\ipykernel_5644\2559747142.py in <module>  
----> 1 car_sales['Price'].plot()  
      2 # price can not run cos is not a numerical data  
  
~\anaconda3\lib\site-packages\pandas\plotting\_core.py in __call__(self, *args, **kwargs)  
    970         data.columns = label_name  
    971  
--> 972         return plot_backend.plot(data, kind=kind, **kwargs)  
    973  
    974     __call__.__doc__ = __doc__  
  
~\anaconda3\lib\site-packages\pandas\plotting\_matplotlib\__init__.py in plot(data, kind, **kwargs)  
    69         kwargs["ax"] = getattr(ax, "left_ax", ax)  
    70     plot_obj = PLOT_CLASSES[kind](data, **kwargs)  
--> 71     plot_obj.generate()  
    72     plot_obj.draw()  
    73     return plot_obj.result  
  
~\anaconda3\lib\site-packages\pandas\plotting\_matplotlib\core.py in generate(self)  
    325     def generate(self):  
    326         self._args_adjust()  
--> 327         self._compute_plot_data()  
    328         self._setup_subplots()  
    329         self._make_plot()  
  
~\anaconda3\lib\site-packages\pandas\plotting\_matplotlib\core.py in _compute_plot_data(self)  
    504         # no non-numeric frames or series allowed  
    505         if is_empty:  
--> 506             raise TypeError("no numeric data to plot")  
    507  
    508         self.data = numeric_data.apply(self._convert_to_ndarray)  
  
TypeError: no numeric data to plot
```

In [70]: 1 car\_sales

Out[70]:

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	\$4,000.00
1	Honda	Red	87899	4	\$5,000.00
2	Toyota	Blue	32549	3	\$7,000.00
3	BMW	Black	11179	5	\$22,000.00
4	Nissan	White	213095	4	\$3,500.00
5	Toyota	Green	99213	4	\$4,500.00
6	Honda	Blue	45698	4	\$7,500.00
7	Honda	Blue	54738	4	\$7,000.00
8	Toyota	White	60000	4	\$6,250.00
9	Nissan	White	31600	4	\$9,700.00

In [71]: 1 car\_sales['Price'].fillna(0)

Out[71]: 0 \$4,000.00  
 1 \$5,000.00  
 2 \$7,000.00  
 3 \$22,000.00  
 4 \$3,500.00  
 5 \$4,500.00  
 6 \$7,500.00  
 7 \$7,000.00  
 8 \$6,250.00  
 9 \$9,700.00

Name: Price, dtype: object

## Convert the price column to a numerical data

```
In [72]: 1 car_sales['Price'] = car_sales['Price'].str.replace('[^\d.]', '') # (clean price column: remove every
2 car_sales['Price'] = car_sales['Price'].astype(float) # convert the price column to float
3 car_sales['Price'] = car_sales['Price'].astype(int) # convert the price column to integer
4 car_sales
```

C:\Users\USER\AppData\Local\Temp\ipykernel\_5644\106475593.py:1: FutureWarning: The default value of regex will change from True to False in a future version.

```
car_sales['Price'] = car_sales['Price'].str.replace('[^\d.]', '') # (clean price column: remove every non-numeric character)
```

Out[72]:

	Make	Colour	Odometer (KM)	Doors	Price
0	Toyota	White	150043	4	4000
1	Honda	Red	87899	4	5000
2	Toyota	Blue	32549	3	7000
3	BMW	Black	11179	5	22000
4	Nissan	White	213095	4	3500
5	Toyota	Green	99213	4	4500
6	Honda	Blue	45698	4	7500
7	Honda	Blue	54738	4	7000
8	Toyota	White	60000	4	6250
9	Nissan	White	31600	4	9700

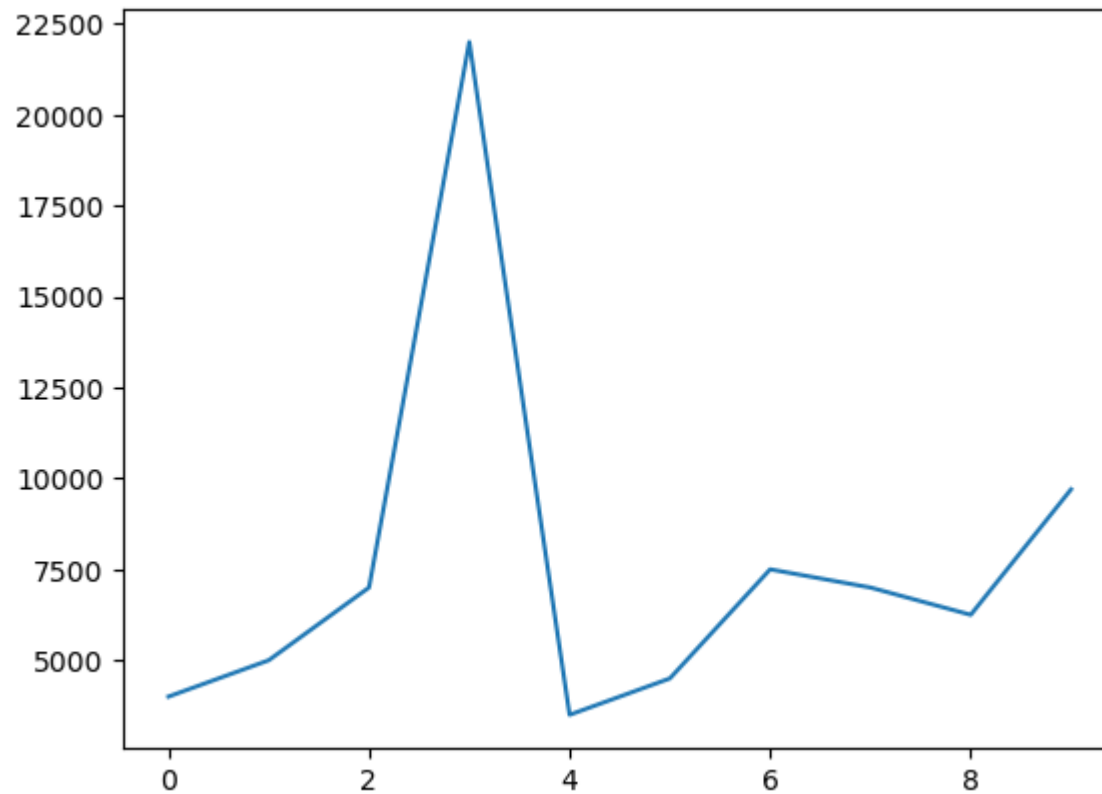
```
In [73]: 1 car_sales['Price'].fillna(0)
```

```
Out[73]: 0      4000  
         1      5000  
         2      7000  
         3     22000  
         4      3500  
         5      4500  
         6      7500  
         7      7000  
         8      6250  
         9      9700  
         Name: Price, dtype: int32
```



```
In [74]: 1 car_sales['Price'].plot()  
        2 # price plot works perfectly
```

Out[74]: <AxesSubplot:>



## Manipulating Data

```
In [75]: 1 car_sales['Make'].str.lower()
```

```
Out[75]: 0    toyota  
1     honda  
2    toyota  
3      bmw  
4    nissan  
5    toyota  
6     honda  
7     honda  
8    toyota  
9    nissan  
Name: Make, dtype: object
```

```
In [76]: 1 car_sales['Make'].str.upper()
```

```
Out[76]: 0    TOYOTA  
1     HONDA  
2    TOYOTA  
3      BMW  
4    NISSAN  
5    TOYOTA  
6     HONDA  
7     HONDA  
8    TOYOTA  
9    NISSAN  
Name: Make, dtype: object
```

```
In [79]: 1 car_sales['Make'] = car_sales['Make'].str.lower()
```

In [80]:

```
1 car_sales
```

Out[80]:

	Make	Colour	Odometer (KM)	Doors	Price
0	toyota	White	150043	4	4000
1	honda	Red	87899	4	5000
2	toyota	Blue	32549	3	7000
3	bmw	Black	11179	5	22000
4	nissan	White	213095	4	3500
5	toyota	Green	99213	4	4500
6	honda	Blue	45698	4	7500
7	honda	Blue	54738	4	7000
8	toyota	White	60000	4	6250
9	nissan	White	31600	4	9700

```
In [81]: 1 car_sales_missing = pd.read_csv('9.1 car-sales-missing-data.csv')
        2 car_sales_missing
```

Out[81]:

	Make	Colour	Odometer	Doors	Price
0	Toyota	White	150043.0	4.0	\$4,000
1	Honda	Red	87899.0	4.0	\$5,000
2	Toyota	Blue	NaN	3.0	\$7,000
3	BMW	Black	11179.0	5.0	\$22,000
4	Nissan	White	213095.0	4.0	\$3,500
5	Toyota	Green	NaN	4.0	\$4,500
6	Honda	NaN	NaN	4.0	\$7,500
7	Honda	Blue	NaN	4.0	NaN
8	Toyota	White	60000.0	NaN	NaN
9	NaN	White	31600.0	4.0	\$9,700

```
In [82]: 1 car_sales_missing['Odometer'].mean()
```

Out[82]: 92302.66666666667

```
In [83]: 1 # how to fill up the missing data
        2 car_sales_missing['Odometer'].fillna(car_sales_missing['Odometer'].mean())
```

Out[83]:

0	150043.000000
1	87899.000000
2	92302.666667
3	11179.000000
4	213095.000000
5	92302.666667
6	92302.666667
7	92302.666667
8	60000.000000
9	31600.000000

Name: Odometer, dtype: float64

In [84]: 1 car\_sales\_missing # *didn't work cos is not assigned*

Out[84]:

	Make	Colour	Odometer	Doors	Price
0	Toyota	White	150043.0	4.0	\$4,000
1	Honda	Red	87899.0	4.0	\$5,000
2	Toyota	Blue	NaN	3.0	\$7,000
3	BMW	Black	11179.0	5.0	\$22,000
4	Nissan	White	213095.0	4.0	\$3,500
5	Toyota	Green	NaN	4.0	\$4,500
6	Honda	NaN	NaN	4.0	\$7,500
7	Honda	Blue	NaN	4.0	NaN
8	Toyota	White	60000.0	NaN	NaN
9	NaN	White	31600.0	4.0	\$9,700

In [85]: 1 # *it can be assigned with the following code:*  
2 # *car\_sales\_missing['Odometer'] = car\_sales\_missing['Odometer'].fillna(car\_sales\_missing['Odometer']).*  
3 # *car\_sales\_missing*

```
In [86]: 1 car_sales_missing['Odometer'].fillna(car_sales_missing['Odometer'].mean(),
2         inplace=True)
3 car_sales_missing
```

Out[86]:

	Make	Colour	Odometer	Doors	Price
0	Toyota	White	150043.000000	4.0	\$4,000
1	Honda	Red	87899.000000	4.0	\$5,000
2	Toyota	Blue	92302.666667	3.0	\$7,000
3	BMW	Black	11179.000000	5.0	\$22,000
4	Nissan	White	213095.000000	4.0	\$3,500
5	Toyota	Green	92302.666667	4.0	\$4,500
6	Honda	NaN	92302.666667	4.0	\$7,500
7	Honda	Blue	92302.666667	4.0	NaN
8	Toyota	White	60000.000000	NaN	NaN
9	NaN	White	31600.000000	4.0	\$9,700

```
In [87]: 1 car_sales_missing.dropna()
```

Out[87]:

	Make	Colour	Odometer	Doors	Price
0	Toyota	White	150043.000000	4.0	\$4,000
1	Honda	Red	87899.000000	4.0	\$5,000
2	Toyota	Blue	92302.666667	3.0	\$7,000
3	BMW	Black	11179.000000	5.0	\$22,000
4	Nissan	White	213095.000000	4.0	\$3,500
5	Toyota	Green	92302.666667	4.0	\$4,500

In [88]: 1 car\_sales\_missing

Out[88]:

	Make	Colour	Odometer	Doors	Price
0	Toyota	White	150043.000000	4.0	\$4,000
1	Honda	Red	87899.000000	4.0	\$5,000
2	Toyota	Blue	92302.666667	3.0	\$7,000
3	BMW	Black	11179.000000	5.0	\$22,000
4	Nissan	White	213095.000000	4.0	\$3,500
5	Toyota	Green	92302.666667	4.0	\$4,500
6	Honda	NaN	92302.666667	4.0	\$7,500
7	Honda	Blue	92302.666667	4.0	NaN
8	Toyota	White	60000.000000	NaN	NaN
9	NaN	White	31600.000000	4.0	\$9,700

In [89]: 1 car\_sales\_missing.dropna(inplace=True)

In [90]: 1 car\_sales\_missing

Out[90]:

	Make	Colour	Odometer	Doors	Price
0	Toyota	White	150043.000000	4.0	\$4,000
1	Honda	Red	87899.000000	4.0	\$5,000
2	Toyota	Blue	92302.666667	3.0	\$7,000
3	BMW	Black	11179.000000	5.0	\$22,000
4	Nissan	White	213095.000000	4.0	\$3,500
5	Toyota	Green	92302.666667	4.0	\$4,500

```
In [91]: 1 # to reaccess the original data, we reimport the data
2 car_sales_missing = pd.read_csv('9.1 car-sales-missing-data.csv')
3 car_sales_missing
```

Out[91]:

	Make	Colour	Odometer	Doors	Price
0	Toyota	White	150043.0	4.0	\$4,000
1	Honda	Red	87899.0	4.0	\$5,000
2	Toyota	Blue	NaN	3.0	\$7,000
3	BMW	Black	11179.0	5.0	\$22,000
4	Nissan	White	213095.0	4.0	\$3,500
5	Toyota	Green	NaN	4.0	\$4,500
6	Honda	NaN	NaN	4.0	\$7,500
7	Honda	Blue	NaN	4.0	NaN
8	Toyota	White	60000.0	NaN	NaN
9	NaN	White	31600.0	4.0	\$9,700

```
In [92]: 1 # creating another data to drop the missing values
2 car_sales_missing_dropped = car_sales_missing.dropna()
3 car_sales_missing_dropped
4
```

Out[92]:

	Make	Colour	Odometer	Doors	Price
0	Toyota	White	150043.0	4.0	\$4,000
1	Honda	Red	87899.0	4.0	\$5,000
3	BMW	Black	11179.0	5.0	\$22,000
4	Nissan	White	213095.0	4.0	\$3,500

```
In [93]: 1 car_sales_missing_dropped.to_csv('car-sales-missing-dropped.csv')
```



# How do we create data from existing data

In [94]:

```
1 car_sales
```

Out[94]:

	Make	Colour	Odometer (KM)	Doors	Price
0	toyota	White	150043	4	4000
1	honda	Red	87899	4	5000
2	toyota	Blue	32549	3	7000
3	bmw	Black	11179	5	22000
4	nissan	White	213095	4	3500
5	toyota	Green	99213	4	4500
6	honda	Blue	45698	4	7500
7	honda	Blue	54738	4	7000
8	toyota	White	60000	4	6250
9	nissan	White	31600	4	9700

In [95]:

```
1 # column from series
2 seats_column = pd.Series([5, 5, 5, 5])
3
4 # new column called seats
5 car_sales['Seats'] = seats_column
6 car_sales
```

Out[95]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats
0	toyota	White	150043	4	4000	5.0
1	honda	Red	87899	4	5000	5.0
2	toyota	Blue	32549	3	7000	5.0
3	bmw	Black	11179	5	22000	5.0
4	nissan	White	213095	4	3500	NaN
5	toyota	Green	99213	4	4500	NaN
6	honda	Blue	45698	4	7500	NaN
7	honda	Blue	54738	4	7000	NaN
8	toyota	White	60000	4	6250	NaN
9	nissan	White	31600	4	9700	NaN

In [96]: 1 car\_sales

Out[96]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats
0	toyota	White	150043	4	4000	5.0
1	honda	Red	87899	4	5000	5.0
2	toyota	Blue	32549	3	7000	5.0
3	bmw	Black	11179	5	22000	5.0
4	nissan	White	213095	4	3500	NaN
5	toyota	Green	99213	4	4500	NaN
6	honda	Blue	45698	4	7500	NaN
7	honda	Blue	54738	4	7000	NaN
8	toyota	White	60000	4	6250	NaN
9	nissan	White	31600	4	9700	NaN

```
In [97]: 1 car_sales['Seats'].fillna(5, inplace=True)
        2 car_sales
```

Out[97]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats
0	toyota	White	150043	4	4000	5.0
1	honda	Red	87899	4	5000	5.0
2	toyota	Blue	32549	3	7000	5.0
3	bmw	Black	11179	5	22000	5.0
4	nissan	White	213095	4	3500	5.0
5	toyota	Green	99213	4	4500	5.0
6	honda	Blue	45698	4	7500	5.0
7	honda	Blue	54738	4	7000	5.0
8	toyota	White	60000	4	6250	5.0
9	nissan	White	31600	4	9700	5.0

In [98]:

```
1 # column from python list
2 fuel_economy = ([7.5, 9.2, 5.0, 9.6, 8.7, 9.0, 9.0, 9.0, 9.0, 9.0])
3 car_sales['fuel per 100KM'] = fuel_economy
4 car_sales
```

Out[98]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM
0	toyota	White	150043	4	4000	5.0	7.5
1	honda	Red	87899	4	5000	5.0	9.2
2	toyota	Blue	32549	3	7000	5.0	5.0
3	bmw	Black	11179	5	22000	5.0	9.6
4	nissan	White	213095	4	3500	5.0	8.7
5	toyota	Green	99213	4	4500	5.0	9.0
6	honda	Blue	45698	4	7500	5.0	9.0
7	honda	Blue	54738	4	7000	5.0	9.0
8	toyota	White	60000	4	6250	5.0	9.0
9	nissan	White	31600	4	9700	5.0	9.0

```
In [99]: 1 car_sales['Total fuel used'] = car_sales['Odometer (KM)']/100 * car_sales['fuel per 100KM']  
2 car_sales
```

Out[99]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM	Total fuel used
0	toyota	White	150043	4	4000	5.0	7.5	11253.225
1	honda	Red	87899	4	5000	5.0	9.2	8086.708
2	toyota	Blue	32549	3	7000	5.0	5.0	1627.450
3	bmw	Black	11179	5	22000	5.0	9.6	1073.184
4	nissan	White	213095	4	3500	5.0	8.7	18539.265
5	toyota	Green	99213	4	4500	5.0	9.0	8929.170
6	honda	Blue	45698	4	7500	5.0	9.0	4112.820
7	honda	Blue	54738	4	7000	5.0	9.0	4926.420
8	toyota	White	60000	4	6250	5.0	9.0	5400.000
9	nissan	White	31600	4	9700	5.0	9.0	2844.000

```
In [100]: 1 # create a column from a single value
          2 car_sales['Number of wheels'] = 4
          3 car_sales
```

Out[100]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM	Total fuel used	Number of wheels
0	toyota	White	150043	4	4000	5.0	7.5	11253.225	4
1	honda	Red	87899	4	5000	5.0	9.2	8086.708	4
2	toyota	Blue	32549	3	7000	5.0	5.0	1627.450	4
3	bmw	Black	11179	5	22000	5.0	9.6	1073.184	4
4	nissan	White	213095	4	3500	5.0	8.7	18539.265	4
5	toyota	Green	99213	4	4500	5.0	9.0	8929.170	4
6	honda	Blue	45698	4	7500	5.0	9.0	4112.820	4
7	honda	Blue	54738	4	7000	5.0	9.0	4926.420	4
8	toyota	White	60000	4	6250	5.0	9.0	5400.000	4
9	nissan	White	31600	4	9700	5.0	9.0	2844.000	4

```
In [101]: 1 car_sales['Passed road safety'] = True
          2 car_sales
```

Out[101]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM	Total fuel used	Number of wheels	Passed road safety
0	toyota	White	150043	4	4000	5.0	7.5	11253.225	4	True
1	honda	Red	87899	4	5000	5.0	9.2	8086.708	4	True
2	toyota	Blue	32549	3	7000	5.0	5.0	1627.450	4	True
3	bmw	Black	11179	5	22000	5.0	9.6	1073.184	4	True
4	nissan	White	213095	4	3500	5.0	8.7	18539.265	4	True
5	toyota	Green	99213	4	4500	5.0	9.0	8929.170	4	True
6	honda	Blue	45698	4	7500	5.0	9.0	4112.820	4	True
7	honda	Blue	54738	4	7000	5.0	9.0	4926.420	4	True
8	toyota	White	60000	4	6250	5.0	9.0	5400.000	4	True
9	nissan	White	31600	4	9700	5.0	9.0	2844.000	4	True

```
In [102]: 1 car_sales.dtypes
```

```
Out[102]: Make                object
          Colour              object
          Odometer (KM)        int64
          Doors                int64
          Price                int32
          Seats                float64
          fuel per 100KM        float64
          Total fuel used       float64
          Number of wheels      int64
          Passed road safety    bool
          dtype: object
```



```
In [103]: 1 # to drop a column  
2 car_sales = car_sales.drop('Total fuel used', axis=1)  
3 car_sales
```

Out[103]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM	Number of wheels	Passed road safety
0	toyota	White	150043	4	4000	5.0	7.5	4	True
1	honda	Red	87899	4	5000	5.0	9.2	4	True
2	toyota	Blue	32549	3	7000	5.0	5.0	4	True
3	bmw	Black	11179	5	22000	5.0	9.6	4	True
4	nissan	White	213095	4	3500	5.0	8.7	4	True
5	toyota	Green	99213	4	4500	5.0	9.0	4	True
6	honda	Blue	45698	4	7500	5.0	9.0	4	True
7	honda	Blue	54738	4	7000	5.0	9.0	4	True
8	toyota	White	60000	4	6250	5.0	9.0	4	True
9	nissan	White	31600	4	9700	5.0	9.0	4	True

In [104]:

```
1 # right way to shuffle your dataframe
2
3 car_sales.sample(frac=1)
```

Out[104]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM	Number of wheels	Passed road safety
8	toyota	White	60000	4	6250	5.0	9.0	4	True
1	honda	Red	87899	4	5000	5.0	9.2	4	True
3	bmw	Black	11179	5	22000	5.0	9.6	4	True
9	nissan	White	31600	4	9700	5.0	9.0	4	True
5	toyota	Green	99213	4	4500	5.0	9.0	4	True
4	nissan	White	213095	4	3500	5.0	8.7	4	True
6	honda	Blue	45698	4	7500	5.0	9.0	4	True
0	toyota	White	150043	4	4000	5.0	7.5	4	True
2	toyota	Blue	32549	3	7000	5.0	5.0	4	True
7	honda	Blue	54738	4	7000	5.0	9.0	4	True

```
In [105]: 1 # to re-assign
          2 car_sales_shuffle = car_sales.sample(frac=1)
          3 car_sales_shuffle
```

Out[105]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM	Number of wheels	Passed road safety
6	honda	Blue	45698	4	7500	5.0	9.0	4	True
0	toyota	White	150043	4	4000	5.0	7.5	4	True
9	nissan	White	31600	4	9700	5.0	9.0	4	True
5	toyota	Green	99213	4	4500	5.0	9.0	4	True
8	toyota	White	60000	4	6250	5.0	9.0	4	True
2	toyota	Blue	32549	3	7000	5.0	5.0	4	True
7	honda	Blue	54738	4	7000	5.0	9.0	4	True
4	nissan	White	213095	4	3500	5.0	8.7	4	True
1	honda	Red	87899	4	5000	5.0	9.2	4	True
3	bmw	Black	11179	5	22000	5.0	9.6	4	True

```
In [106]: 1 # only select 20% of data
          2 car_sales_shuffle.sample(frac=0.2)
```

Out[106]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM	Number of wheels	Passed road safety
2	toyota	Blue	32549	3	7000	5.0	5.0	4	True
3	bmw	Black	11179	5	22000	5.0	9.6	4	True

```
In [107]: 1 car_sales_shuffle.sample(frac=0.1)
```

Out[107]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM	Number of wheels	Passed road safety
1	honda	Red	87899	4	5000	5.0	9.2	4	True

In [108]: 1 car\_sales\_shuffle.sample(frac=0.01)

Out[108]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM	Number of wheels	Passed road safety
--	------	--------	---------------	-------	-------	-------	----------------	------------------	--------------------

In [109]: 1 car\_sales\_shuffle

Out[109]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM	Number of wheels	Passed road safety
6	honda	Blue	45698	4	7500	5.0	9.0	4	True
0	toyota	White	150043	4	4000	5.0	7.5	4	True
9	nissan	White	31600	4	9700	5.0	9.0	4	True
5	toyota	Green	99213	4	4500	5.0	9.0	4	True
8	toyota	White	60000	4	6250	5.0	9.0	4	True
2	toyota	Blue	32549	3	7000	5.0	5.0	4	True
7	honda	Blue	54738	4	7000	5.0	9.0	4	True
4	nissan	White	213095	4	3500	5.0	8.7	4	True
1	honda	Red	87899	4	5000	5.0	9.2	4	True
3	bmw	Black	11179	5	22000	5.0	9.6	4	True

In [110]: 1 car\_sales\_shuffle.reset\_index()

Out[110]:

	index	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM	Number of wheels	Passed road safety
0	6	honda	Blue	45698	4	7500	5.0	9.0	4	True
1	0	toyota	White	150043	4	4000	5.0	7.5	4	True
2	9	nissan	White	31600	4	9700	5.0	9.0	4	True
3	5	toyota	Green	99213	4	4500	5.0	9.0	4	True
4	8	toyota	White	60000	4	6250	5.0	9.0	4	True
5	2	toyota	Blue	32549	3	7000	5.0	5.0	4	True
6	7	honda	Blue	54738	4	7000	5.0	9.0	4	True
7	4	nissan	White	213095	4	3500	5.0	8.7	4	True
8	1	honda	Red	87899	4	5000	5.0	9.2	4	True
9	3	bmw	Black	11179	5	22000	5.0	9.6	4	True

```
In [111]: 1 car_sales_shuffle.reset_index(drop = True, inplace=True)
          2 car_sales_shuffle
```

Out[111]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM	Number of wheels	Passed road safety
0	honda	Blue	45698	4	7500	5.0	9.0	4	True
1	toyota	White	150043	4	4000	5.0	7.5	4	True
2	nissan	White	31600	4	9700	5.0	9.0	4	True
3	toyota	Green	99213	4	4500	5.0	9.0	4	True
4	toyota	White	60000	4	6250	5.0	9.0	4	True
5	toyota	Blue	32549	3	7000	5.0	5.0	4	True
6	honda	Blue	54738	4	7000	5.0	9.0	4	True
7	nissan	White	213095	4	3500	5.0	8.7	4	True
8	honda	Red	87899	4	5000	5.0	9.2	4	True
9	bmw	Black	11179	5	22000	5.0	9.6	4	True

```
In [112]: 1 # apply a function to a column  
2 car_sales
```

Out[112]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM	Number of wheels	Passed road safety
0	toyota	White	150043	4	4000	5.0	7.5	4	True
1	honda	Red	87899	4	5000	5.0	9.2	4	True
2	toyota	Blue	32549	3	7000	5.0	5.0	4	True
3	bmw	Black	11179	5	22000	5.0	9.6	4	True
4	nissan	White	213095	4	3500	5.0	8.7	4	True
5	toyota	Green	99213	4	4500	5.0	9.0	4	True
6	honda	Blue	45698	4	7500	5.0	9.0	4	True
7	honda	Blue	54738	4	7000	5.0	9.0	4	True
8	toyota	White	60000	4	6250	5.0	9.0	4	True
9	nissan	White	31600	4	9700	5.0	9.0	4	True

```
In [113]: 1 car_sales['Odometer (KM)'] = car_sales['Odometer (KM)'].apply(lambda x: x / 1.6)
          2 car_sales
```

Out[113]:

	Make	Colour	Odometer (KM)	Doors	Price	Seats	fuel per 100KM	Number of wheels	Passed road safety
0	toyota	White	93776.875	4	4000	5.0	7.5	4	True
1	honda	Red	54936.875	4	5000	5.0	9.2	4	True
2	toyota	Blue	20343.125	3	7000	5.0	5.0	4	True
3	bmw	Black	6986.875	5	22000	5.0	9.6	4	True
4	nissan	White	133184.375	4	3500	5.0	8.7	4	True
5	toyota	Green	62008.125	4	4500	5.0	9.0	4	True
6	honda	Blue	28561.250	4	7500	5.0	9.0	4	True
7	honda	Blue	34211.250	4	7000	5.0	9.0	4	True
8	toyota	White	37500.000	4	6250	5.0	9.0	4	True
9	nissan	White	19750.000	4	9700	5.0	9.0	4	True

In [ ]:

1